

# PVS Linear Algebra Libraries for Verification of Control Software Algorithms

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## Outline

- 1 Introduction
- 2 Linear Map
- 3 Matrices
- 4 Invertibility and Isomorphisms
- 5 Control Theory
- 6 Control Theory Verification
- 7 Conclusions

- The objective of control theory is to calculate a proper action from the controller that will result in stability for the system
- The software implementation of a control law can be inspected by analysis tools
- However these tools are often challenged by issues for which solutions are already available from control theory.

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- Program verification uses proof assistants to ensure the validity of user-provided code annotations.
- These annotations may express the domain-specific properties of the code.
- However, formulating annotations correctly is nontrivial in practice.
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- 1 Axiomatization of Lyapunov-based stability as C code annotations,
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- If there exists a positive definite function  $V$  such that  $V(\xi(k)) \leq 1$  implies  $V(\xi(k+1)) \leq 1$  then this function can be used to establish the stability of the system.
- This Lyapunov function,  $V$ , defines the ellipsoid  $\{\xi \mid V(\xi) \leq 1\}$ , this ellipsoid plays an important role for the stability preservation at the code level.

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Annotated with assertions in the Hoare style we get



$$\begin{array}{c} \{pre1\} \\ u = C_c \mathbf{x}_c + D_c y_c \\ \{post1\} \end{array}$$



$$\begin{array}{c} \{pre2\} \\ \mathbf{x}_c = A_c \mathbf{x}_c + B_c y_c \\ \{post2\}. \end{array}$$

- To use ellipsoids to formally specify bounded input, bounded state.
- Typically, an instruction  $S$  would be annotated in the following way:

$$\{x \in \mathcal{E}_P\} y = Ax + b \{y - b \in \mathcal{E}_Q\} \quad (1)$$

where the pre- and post- conditions are predicates expressing that the variables belong to some ellipsoid, with  $\mathcal{E}_P = \{x : \mathbb{R}^n | x^T P^{-1} x \leq 1\}$  and  $Q = APA^T$ .



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## An ellipsoid-aware Hoare logic

The mathematical theorem that guarantees the relations is :

### Theorem

*If  $M, Q$  are invertible matrices, and*

$$(x - c)^T Q^{-1} (x - c) \leq 1 \text{ and}$$

$$y = Mx + b$$

*then*

$$(y - b - Mc)^T (MQM^T)^{-1} (y - b - Mc) \leq 1$$

We will refer to it as the *ellipsoid theorem*.

- The pre- and post- conditions are expressed as predicates in ACSI and PVS.
- The multiplication of a matrix with a vector is defined with function `vect_mult(matrix  $A$ , vector  $x$ )`, which returns a vector.
- Addition and multiplication of 2 matrices, multiplication by a scalar, and inverse of a matrix are declared as matrix types

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inverse of a matrix  $A$ ,  $\text{mat\_inverse}(A)$  is defined using the predicate  $\text{is\_invertible}(A)$  as follows:

```
/*@ axiom mat_inv_select_i_eq_j:
@  ∀matrix A, integer i, j;
@  is_invertible(A) && i == j ==>
@  mat_select(mat_mult(A, mat_inverse(A)), i, j) = 1
@
@ axiom mat_inv_select_i_dff_j:
@  ∀matrix A, integer i, j;
@  is_invertible(A) && i != j ==>
@  mat_select(mat_mult(A, mat_inverse(A)), i, j) = 0
@*/
```

ACSL

Complex constructions or relations can be defined as uninterpreted predicates. The following predicate is meant to express that vector  $x$  belongs to  $\mathcal{E}_{\mathcal{P}}$ :

```
● // @ predicate in_ellipsoid(matrix P, vector x);
```

ACSL

- The paramount notion in ACSL is the function contract.
- The key word `requires` is used to introduce the pre-conditions of the triple, and the key word `ensures` is used to introduce its post-conditions.
- `//@ requires P`  
`//@ ensures R`  
`Q`

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```
in_ellipsoid?(P_0, vect_of_array(xc, 2, floatP_floatM))))))
IMPLIES
in_ellipsoid?(Q, vect_of_array(yc, 2, floatP_floatM0))
```

pvs

```
vect_of_array(yc, 2, floatP_floatM0)'vect =
Ac * vect_of_array(xc, 2, floatP_floatM)'vect
```

pvs

For both POs,

- we must first interpret the uninterpreted types and to prove the properties that are defined axiomatically.
- We must then discharge the verification conditions. This is done by using PVS and a linear algebra extension of it.

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For both POs,

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In order to discharge the PO, the following libraries need to be used:

- Linear\_algebra:
  - linear\_map, matrices, matrix\_operator, block\_matrices
- Control\_theory
  - ellipsoid, s\_procedure\_def, shur\_formula

## linear\_map

$$T : [n, m, Vector[n] \rightarrow Vector[m]]$$

Mapping:TYPE= [# dom: posnat, codom: posnat, mp:  
[Vector[dom]->Vector[codom]] #]

pvs

$$(f, g) \longmapsto [n, m, f' mp(x) + g' mp(x)]$$

+(f: Mapping, (g: (same\_dim?(f)))): Mapping =  
f WITH ['mp:= lambda(x: Vector[f'dom]): f' mp(x) + g' mp(x)]

pvs

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```

pvs

## linear\_map\_def

$$\{e(i) \in \mathbb{R}^n | i = 1, \dots, n\}$$

```
unit?(n)(e: [below[n] -> Vector[n]]):
bool = FORALL (i: below[n]): e(i)*e(i)=1
ortho?(n)(e: [below[n] -> Vector[n]]):
bool = FORALL (i,j: below[n]): (i /= j IMPLIES e(i)*e(j)=0)
```

pvs

$$x = \sum_{i=1}^n (x e(i)) e(i)$$

```
vec_expan?(n)(e: [below[n] -> Vector[n]]):
bool = FORALL (x: Vector[n]): x = SigmaV(0,n-1,x*e)
```

pvs



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$$x = \sum_{i=1}^n (xe(i))e(i)$$

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## linear\_map\_def

$$\{e(i) \in \mathbb{R}^n | i = 1, \dots, n\}$$

```
base?(n)(e: [below[n] -> Vector[n]]): bool = unit?(n)(e) and
ortho?(n)(e) and vec_expan?(n)(e)
```

pvs

```
e(n): [below[n]->Vector[n]] = LAMBDA(j: below[n]): LAMBDA(i:
below[n]): IF (i=j) THEN 1 ELSE 0 ENDIF
```

exercise

```
cano_base:LEMMA base?(n)(e(n))
```

exercise

## linear\_map\_def

$$\{e(i) \in \mathbb{R}^n | i = 1, \dots, n\}$$

base?(n)(e: [below[n] -> Vector[n]]): bool = unit?(n)(e) and  
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exercise

## linear\_map\_def

$$h\left[\sum_{i=0}^{l-1} x_i F_i\right] = \sum_{i=0}^{l-1} h[x_i F_i]$$

linear\_map\_e?(h,l,n,m): bool = h'dom = n and h'codom = m and  
FORALL (x: Vector[l],F: [below[l]->Vector[n]]):  
h'mp(SigmaV[below[l],n](0,l-1,x\*F)) = SigmaV[below[l],m](0,l -  
1,x\*(h'mp o F));

pvs

$$\begin{aligned} h[x + y] &= h[x] + h[y] \\ h[ax] &= ah[x] \end{aligned}$$

additive?(f): bool = FORALL (x,y: Vector[f'dom]):  
f'mp(x + y) = f'mp(x) + f'mp(y)  
homogeneous?(f): bool = FORALL (a: real, x: Vector[f'dom]):  
f'mp(a\*x) = a\*f'mp(x)  
linear\_map?(f): bool = additive?(f) AND homogeneous?(f)

pvs

# linear\_map\_def

```
linear_map_characterization: LEMMA FORALL (f: Map(n,n)):
linear_map?(f) IFF linear_map_e?(n,n)(f)
```

exercise

# matrices

$$A : [n, m, (i, j) \mapsto A_{i,j}]$$

```
Matrix: TYPE = [# rows: posnat, cols: posnat,
matrix: [below(rows), below(cols) -> real] #]
```

pvs

$$+ : (M, N) \mapsto [n, m, (i, j) \mapsto M_{i,j} + N_{i,j}]$$

```
+(M, (N: (same_dim?(M)))): Matrix = M WITH [ 'matrix :=
LAMBDA (i: below(M'rows), j: below(M'cols)):
M'matrix(i, j) + N'matrix(i, j) ];
```

pvs

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pvs

# matrices

$$M * M^{-1} = M^{-1} * M = 1$$

inverse?(M: Square)(N: Square | N'rows = M'rows): bool =  
M \* N = I(M'rows) and N \* M = I(M'rows)

pvs

invertible?(M: Square): bool = EXISTS (N: (inverse?(M))):  
inverse?(M) (N)

pvs

inverse\_unique: lemma FORALL (M: (invertible?), N, P:  
(inverse?(M))): N = P

exercise

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pvs

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inverse_unique: lemma FORALL (M: (invertible?), N, P:
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```

exercise

# matrix\_operator

$$L(n, m) : f \longmapsto [m, n, (i, j) \longmapsto f(e(n)(i))(j) = A_{ij}]$$

```
L(n,m): [Map_linear(n,m) -> Mat(m,n)] =
  (lambda(f: Map_linear(n,m))):
  (# rows:= m,cols:= n,matrix:=
  lambda(j: below[m],i: below[n]): f'mp(e(n)(i))(j)#))
```

pvs

$$T(n, m) : A \longmapsto [n, m, x \longmapsto A * x]$$

```
T(n,m): [Mat(m,n) -> Map_linear(n,m)]=
  (lambda(A: Mat(m,n)): (#dom:= n,codom:= m,mp:=
  lambda(x: Vector[n]): lambda(j: below[m]):
  sigma(0,A'cols-1,lambda(i: below[A'cols]):
  A'matrix(j,i)*x(i))#))
```

pvs

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```

pvs

## matrix\_operator

```
Iso: LEMMA bijective?(L(n,m))
```

exercise

```
map_matrix_bij: LEMMA FORALL (A: Mat(m,n)):
  L(n,m)(T(n,m)(A)) = A
iso_map: LEMMA FORALL (f: Map_linear(n,m)):
  T(n,m)(L(n,m)(f)) = f
```

exercise

# matrix\_operator

comp\_mult: LEMMA FORALL (g: Map\_linear(n,m),f: Map\_linear(m,p)):  
 $L(n,p)(f \circ g) = L(m,p)(f) * L(n,m)(g)$  exercise

Matrix\_inv(n):type= {A: Square | squareMat?(n)(A) and  
 bijective?(n)(T(n,n)(A))} pvs

$$\begin{array}{ccc} inv(n) : Matrix\_inv(n) & \longrightarrow & Matrix\_inv(n) \\ A & \longmapsto & L_{n,n}((T_{n,n}(A))^{-1}) \end{array}$$

$inv(n)(A) = L(n,n)(inverse(n)(T(n,n)(A)))$  pvs

# matrix\_operator

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 $L(n,p)(f \circ g) = L(m,p)(f) * L(n,m)(g)$  exercise

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$inv(n)(A) = L(n,n)(inverse(n)(T(n,n)(A)))$  pvs

How is the  $inverse?(M)$ ?

```
inv: LEMMA squareMat?(n)(M) AND bijective?(n)(T(n,n)(M))
      IMPLIES inverse?(M)(inv(n)(M))
```

exercise

## matrix\_operator

```
prod_inv_oper: LEMMA square?(A) and squareMat?(n)(A) and
      bijective?(n)(T(n,n)(A)) AND
      square?(B) and squareMat?(n)(B) and bijective?(n)(T(n,n)(B))
      IMPLIES
      inv(n)(A*B)=inv(n)(B)*inv(n)(A)
```

exercise



$$M = \begin{pmatrix} M_1 & M_3 \\ M_2 & M_4 \end{pmatrix}$$

$$M : [row1, row2, cols1, cols2, (i, j) \mapsto M_{i,j}] \quad (2)$$

Block\_Matrix: TYPE = [# rows1: posnat, rows2: posnat,  
cols1: posnat, cols2: posnat,  
matrix: [below(rows1 + rows2), below(cols1 + cols2) -> real] #]

pvs

$$Block2M1 : M \mapsto [rows1, cols1, (i, j) \mapsto M_{i,j}]$$

Block2M1(M): Matrix = (# rows := M'rows1, cols := M'cols1,  
matrix := LAMBDA (i: below(M'rows1), j: below(M'cols1)):  
M'matrix(i,j) #)

pvs

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## schur\_formula

$$M = \begin{pmatrix} M_1 & M_3 \\ M_2 & M_4 \end{pmatrix} > 0 \quad (3)$$

$$M_4 > 0 \text{ and } (M_1 - M_3(M_4)^{-1}M_2) > 0 \quad (4)$$

strict\_schur\_formula: LEMMA (Block2M2(M) =  
 transpose(Block2M3(M)) AND invertible?(Block2M4(M)) AND  
 symmetric?(Block2M1(M)) AND symmetric?(Block2M4(M)))  
 IMPLIES  
 (def\_pos?(M)  
 IFF  
 def\_pos?(Block2M4(M)) AND def\_pos?(Block2M1(M) -  
 Block2M3(M)\*inverse(Block2M4(M))\*Block2M2(M)))

pvs

## S-procedure

Thus the implication that must be proved is as follows:

$$\{\mathbf{x}^T P \mathbf{x} \leq 1, \text{ and } y^2 \leq 1\} \text{ implies } (A\mathbf{x} + By)^T P (A\mathbf{x} + By) \leq 1. \quad (5)$$

Applying the S-procedure the implication 5 is equivalent to: Exist a  $\mu \in \mathbb{R}$

$$(A\mathbf{x} + By)^T P (A\mathbf{x} + By) - \mu \mathbf{x}^T P \mathbf{x} - (1 - \mu)y^2 \leq 0.$$

# S-procedure

## Control\_theory

Let the linear functionals  $\sigma_k : R^n \rightarrow R$  and consider the following two conditions

- 1  $S_1$ : For all  $k = 1, 2, \dots, N$ ,  $\sigma_k > 0$  implies  $\sigma_0 \geq 0$

`s1_condition?(m)(beta: fun_constraint(m), f: Map(n,1)): bool`  
`= FORALL (x: Vector[n]): pos_constraint_point?(m)(beta,x)`  
`IMPLIES f'mp(x)(0) >= 0`

- 2  $S_2$ : There exists  $\tau_k \geq 0$ ,  $k = 1, 2, \dots, N$  such that

$$\sigma_0(y) - \sum_{k=1}^N \tau_k \sigma_k(y) \geq 0, \forall y \in R^n$$

`s2_condition?(m)(beta: fun_constraint(m), f: Map(n,1)): bool`  
`= EXISTS (r: pos_scalar_family(m)): (FORALL (x: Vector[n]):`  
`f'mp(x)(0) - sigma[below[m]](0,m - 1, LAMBDA(i: below[m]):`  
`r(i)*beta(i)'mp(x)(0)) >= 0)`

## Control\_theory

`ellipsoid: LEMMA  $\forall$  (n:posnat, Q, M: SquareMat(n), x, y, b, c:`  
`Vector[n]):`  
`bijective?(n)(T(n,n)(Q)) AND bijective?(n)(T(n,n)(M))`  
`AND (x-c)*(inv(n)(Q)*(x-c))  $\leq$  1`  
`AND y=M*x + b`  
`IMPLIES`  
`(y-b-M*c)*(inv(n)(M*(Q*transpose(M)))*(y-b-M*c))  $\leq$  1`

# Control\_theory\_verification

$$\{x \in \mathcal{E}_P\} y = Mx + b \{y - b \in \mathcal{E}_Q\} \quad (6)$$

• in\_ellipsoid?(P, X) and Y=MX+b  
IMPLIES  
in\_ellipsoid?(MQM<sup>T</sup>, Y-b)

pvs

• bijections :LEMMA  
bijective?(2)(T(2,2)(Q) AND bijective?(2)(T(2,2)(M))

pvs

# Control\_theory\_verification

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# Conclusions

- We have outlined a global approach to validate stability properties of code implementing controllers.
- Our approach requires the code to be annotated by Hoare triples,
- Linear algebra PVS libraries can be used for the formal specification of control theory properties
- We have defined a PVS library able to manipulate predicates over the code.

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# Conclusions

- We have outlined a global approach to validate stability properties of code implementing controllers.
- Our approach requires the code to be annotated by Hoare triples,
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- Our approach requires the code to be annotated by Hoare triples,
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- Linear Algebra:
  - `sigma_lemmas`, `linear_map`, `sigma_vector`, `linear_map_def`, `vect_of_vect`
  - `matrices`, `matrix_operator`, `matrix_lemmas`, `block_matrices`
- Control Theory:
  - `ellipsoid`, `convex_def`, `s_procedure_def`
  - `schur_prelim`, `schur_formula`