

Translation of Lyapunov Stability Proofs to Machine Arithmetic

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Eighth meeting of the French community of compilation

Nice, July 2014

Embedded Systems

An **embedded system** is a computer system with a dedicated function, within a larger mechanical or electrical system.

Constraints:

- Power consumption;
- Performance (RT);
- Safety;
- Cost.

Uses a low-power processor or a microcontroller.

Commonly found in consumer, cooking, industrial, automotive, medical, commercial and military applications.

Example

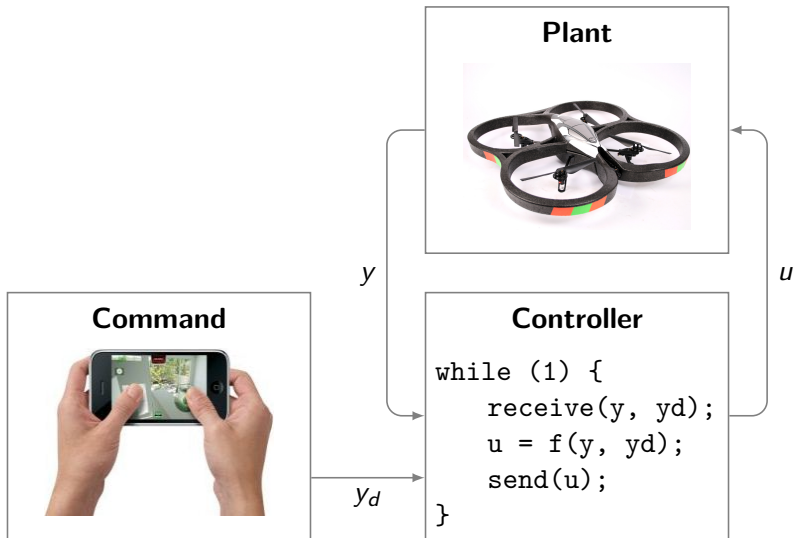
Quadricopter, DRONE Project, MINES ParisTech & ÉCP
⇒ Parrot AR.Drone.



ATMEGA128: 16 MHz, 4 KB RAM, 128 KB ROM



Control-Command System



Levels of Description

Formalization:

- System conception;
- Constraint specification;
- Physical model of the environment;
- Mathematical **proof** that the system behave properly.

MATLAB, Simulink

Realization: very low-level C program

- Thousands of LOC;
- Computations decomposed into elementary operations;
- Management of sensors and actuators.

GCC, Clang



Gradual **transformations**

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Formalization:

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- Constraint specification;
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- Computations decomposed into elementary operations;
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Gradual **transformations**



How to ensure that the executed program is correct?

Levels of Description

Formalization:

-
-
-
-

MAT

Realization: very low-level C



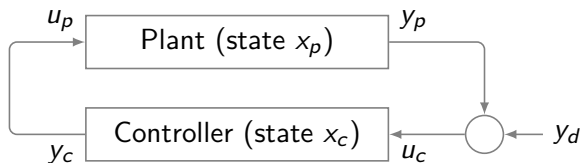
is;
ors

How to ensure that the executed program is correct?

Stability Proof

Show that the system parameters are **bounded** during its execution.

Essential for system safety.



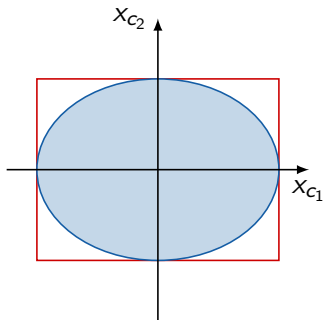
- **Open loop stability:** u_c bounded $\implies x_c$ bounded
(hence y_c bounded)
- **Closed loop stability:** y_d bounded $\implies x_c, x_p$ bounded
(hence y_c, y_p bounded)

Stability Invariant

Lyapunov theory provides a framework to compute inductive invariants.

Linear invariants not well suited.

Quadratic invariants (ellipsoids) are a good fit for linear systems.



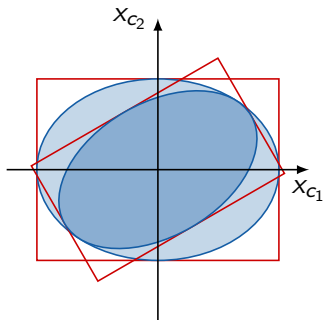
Static analysis to show that the invariant holds from source code.

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Numerical Precision

Lyapunov theory applies on a system with **real arithmetic**.

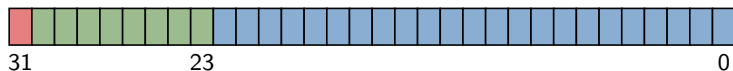
In machine implementations, numerical values are **approximated** by binary, **limited-precision** values.

Numerical Precision

Lyapunov theory applies on a system with **real arithmetic**.

In machine implementations, numerical values are **approximated** by binary, **limited-precision** values.

- **Floating point** (IEEE 754):



$$(-1)^s \times 2^{e-127} \times m$$

- **Fixed point**:

$$(-1)^s \times e + 2^{-24} \times m$$

- **Rationals** using pairs of integers.

Numerical Precision

Lyapunov theory applies on a system with **real arithmetic**.

In machine implementations, numerical values are **approximated** by binary, **limited-precision** values.

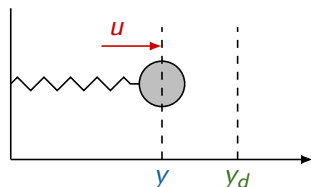
- 1 Constant values are altered;
- 2 Rounding errors during computations.

⇒ Stability proof does not apply, invariant does not fit.

How to adapt the stability proof?

Example System

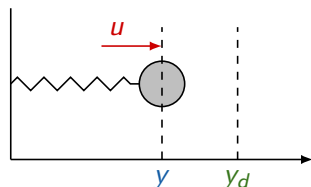
[Feron ICSM'10]:
mass-spring system.



```
Ac = [0.4990, -0.0500;  
      0.0100, 1.0000];  
Bc = [1; 0];  
Cc = [564.48, 0];  
Dc = -1280;  
xc = zeros(2, 1);  
receive(y, 2); receive(yd, 3);  
while (1)  
    yc = max(min(y - yd, 1), -1);  
    u = Cc*xc + Dc*yc;  
    xc = Ac*xc + Bc*yc;  
    send(u, 1);  
    receive(y, 2); receive(yd, 3);  
end
```

Example System

[Feron ICSM'10]:
mass-spring system.



Open-loop stability:
 x_c bounded.

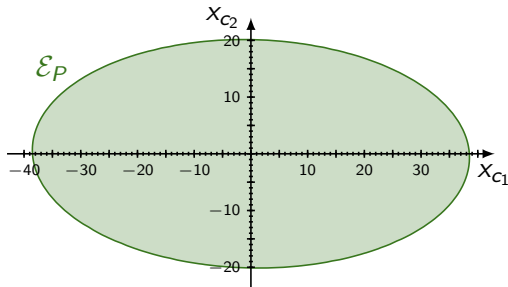
```
Ac = [0.4990, -0.0500;  
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    u = Cc*xc + Dc*yc;  
    xc = Ac*xc + Bc*yc;  
    send(u, 1);  
    receive(y, 2); receive(yd, 3);  
end
```

Example System: Stability Ellipse

Lyapunov theory $\implies x_c = \begin{pmatrix} x_{c1} \\ x_{c2} \end{pmatrix}$ belongs to the ellipse:

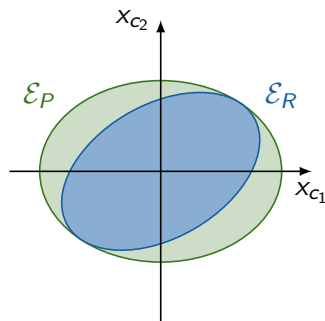
$$\mathcal{E}_P = \{x \in \mathbb{R}^2 \mid x^T \cdot P \cdot x \leq 1\} \quad P = 10^{-3} \begin{pmatrix} 0.6742 & 0.0428 \\ 0.0428 & 2.4651 \end{pmatrix}$$

$$x_c \in \mathcal{E}_P \iff 0.6742x_{c1}^2 + 0.0856x_{c1}x_{c2} + 2.4651x_{c2}^2 \leq 1000$$



Example System

```
Ac = [0.4990, -0.0500;  
      0.0100, 1.0000];  
Bc = [1; 0];  
Cc = [564.48, 0];  
Dc = -1280;  
xc = zeros(2, 1);  
receive(y, 2); receive(yd, 3);  
while (1)  
    %  $x_c \in \mathcal{E}_P$   
    yc = max(min(y - yd, 1), -1);  
    u = Cc*xc + Dc*yc;  
    xc = Ac*xc + Bc*yc;  
    send(u, 1);  
    receive(y, 2); receive(yd, 3);  
    %  $x_c \in \mathcal{E}_R \subset \mathcal{E}_P$   
end
```



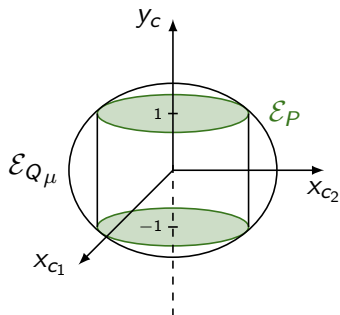
Example System: Invariants

```
xc = zeros(2, 1);  
%  $x_c \in \mathcal{E}_P$   
receive(y, 2); receive(yd, 3);  
%  $x_c \in \mathcal{E}_P$   
while (1)  
    %  $x_c \in \mathcal{E}_P$   
    yc = max(min(y - yd, 1), -1);  
    %  $x_c \in \mathcal{E}_P, y_c^2 \leq 1$   
    %  $\begin{pmatrix} x_c \\ y_c \end{pmatrix} \in \mathcal{E}_{Q_\mu}, Q_\mu = \begin{pmatrix} \mu P & 0 \\ 0 & 1 - \mu \end{pmatrix}, \mu = 0.9991$   
    u = Cc*xc + Dc*yc;  
    %  $\begin{pmatrix} x_c \\ y_c \end{pmatrix} \in \mathcal{E}_{Q_\mu}$   
    xc = Ac*xc + Bc*yc;  
    %  $x_c \in \mathcal{E}_R, R = [(A_c \ B_c)Q_\mu^{-1}(A_c \ B_c)^T]^{-1}$   
    send(u, 1);  
    %  $x_c \in \mathcal{E}_R$   
    receive(y, 2); receive(yd, 3);  
    %  $x_c \in \mathcal{E}_R$   
    %  $x_c \in \mathcal{E}_P$   
end
```

Example System: Invariants

$$\% x_c \in \mathcal{E}_P, \quad y_c^2 \leq 1$$

$$\% \begin{pmatrix} x_c \\ y_c \end{pmatrix} \in \mathcal{E}_{Q_\mu}, \quad Q_\mu = \begin{pmatrix} \mu^P & 0 \\ 0 & 1-\mu \end{pmatrix}, \quad \mu = 0.9991$$



$$\% \begin{pmatrix} x_c \\ y_c \end{pmatrix} \in \mathcal{E}_{Q_\mu}$$

$$\mathbf{x}_c = \mathbf{A}_c \mathbf{x}_c + \mathbf{B}_c \mathbf{y}_c;$$

$$\% x_c \in \mathcal{E}_R, \quad R = [(\mathbf{A}_c \ \mathbf{B}_c) \mathbf{Q}_\mu^{-1} (\mathbf{A}_c \ \mathbf{B}_c)^T]^{-1}$$

Example System

```
Ac = [0.4990, -0.0500;  
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    send(u, 1);  
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end
```

Using limited-precision
arithmetic:

Example System

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Ac = [0.4990, -0.0500;  
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Using limited-precision arithmetic:

- 1 Constant values are altered

Example System

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    xc = Ac*xc + Bc*yc;  
    send(u, 1);  
    receive(y, 2); receive(yd, 3);  
    %  $x_c \in \mathcal{E}_P$   
end
```

Using limited-precision arithmetic:

- 1 Constant values are altered
 $\implies \mathcal{E}_P$ no longer valid;

Example System

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Ac = [0.4990, -0.0500;  
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Dc = -1280;  
xc = zeros(2, 1);  
receive(y, 2); receive(yd, 3);  
while (1)  
    %  $x_e \in \mathcal{E}_P$   
    yc = max(min(y - yd, 1), -1);  
    u = Cc*xc + Dc*yc;  
    xc = Ac*xc + Bc*yc;  
    send(u, 1);  
    receive(y, 2); receive(yd, 3);  
    %  $x_e \in \mathcal{E}_P$   
end
```

Using limited-precision arithmetic:

- 1 Constant values are altered
 $\implies \mathcal{E}_P$ no longer valid;
- 2 Rounding errors during computations.

Example System

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Ac = [0.4990, -0.0500;  
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    send(u, 1);  
    receive(y, 2); receive(yd, 3);  
    %  $x_c \in \mathcal{E}_P$   
end
```

Using limited-precision arithmetic:

- 1 Constant values are altered
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- 2 Rounding errors during computations.

Adapt invariants.

Theoretical Framework

Transpose code + invariants in two steps:

Real

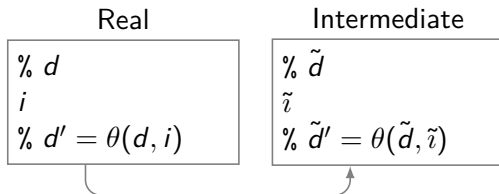
```
% d
```

```
i
```

```
%  $d' = \theta(d, i)$ 
```

Theoretical Framework

Transpose code + invariants in two steps:

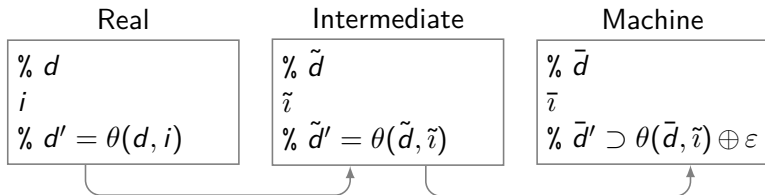


Code: constants converted into machine numbers

Invariants recomputed using the same propagation theorem θ

Theoretical Framework

Transpose code + invariants in two steps:



Code: constants converted into machine numbers

Invariants recomputed using the same propagation theorem θ

Code: real functions +, *... replaced by their machine counterparts

Invariants enlarged to include rounding error
Preserve invariant shape for propagation

Example System, 32-bit Floating-Point Numbers

```
xc = zeros(2, 1);  
%  $x_c \in \mathcal{E}_P$   
receive(y, 2); receive(yd, 3);  
%  $x_c \in \mathcal{E}_P$   
while (1)  
    %  $x_c \in \mathcal{E}_P$   
    yc = max(min(y - yd, 1), -1);  
    %  $x_c \in \mathcal{E}_P, y_c^2 \leq 1$   
    %  $\begin{pmatrix} x_c \\ y_c \end{pmatrix} \in \mathcal{E}_{Q_\mu}, Q_\mu = \begin{pmatrix} \mu^P & 0 \\ 0 & 1-\mu \end{pmatrix}$   
    u = Cc*xc + Dc*yc;  
    %  $\begin{pmatrix} x_c \\ y_c \end{pmatrix} \in \mathcal{E}_{Q_\mu}$   
    xc = Ac*xc + Bc*yc;  
    %  $x_c \in \mathcal{E}_R, R = [(A_c \ B_c)Q_\mu^{-1}(A_c \ B_c)^T]^{-1}$   
    send(u, 1);  
    %  $x_c \in \mathcal{E}_R$   
    receive(y, 2); receive(yd, 3);  
    %  $x_c \in \mathcal{E}_R$   
    %  $x_c \in \mathcal{E}_P$   
end
```

In the rest of the code:

Example System, 32-bit Floating-Point Numbers

```
xc = zeros(2, 1);  
% xc ∈ EP  
receive(y, 2); receive(yd, 3);  
% xc ∈ EP  
while (1)  
    % xc ∈ EP  
    yc = max(min(y - yd, 1), -1);  
    % xc ∈ EP, yc2 ≤ 1  
    % (xc / yc) ∈ EQμ, Qμ = ( μP 0 / 0 1-μ )  
    u = Cc*xc + Dc*yc;  
    % (xc / yc) ∈ EQμ  
    xc = Acf*xc + Bcf*yc;  
    % xc ∈ ER, R = [(Ac Bc)Qμ-1(Ac Bc)T]-1  
    send(u, 1);  
    % xc ∈ ER  
    receive(y, 2); receive(yd, 3);  
    % xc ∈ ER  
    % xc ∈ EP  
end
```

In the rest of the code:

- A_c, B_c replaced by A_{cf}, B_{cf} ;

Example System, 32-bit Floating-Point Numbers

```
xc = zeros(2, 1);
% xc ∈ EP
receive(y, 2); receive(yd, 3);
% xc ∈ EP
while (1)
    % xc ∈ EP
    yc = max(min(y - yd, 1), -1);
    % xc ∈ EP, yc2 ≤ 1
    % (xc / yc) ∈ EQμ, Qμ = ( μP 0 / 0 1-μ )
    u = Cc*xc + Dc*yc;
    % (xc / yc) ∈ EQμ
    xc = Acf*xc + Bcf*yc;
    % xc ∈ ES, S = [(Acf Bcf)Qμ-1(Acf Bcf)T]-1
    send(u, 1);
    % xc ∈ ES
    receive(y, 2); receive(yd, 3);
    % xc ∈ ES
    % xc ∈ EP
end
```

In the rest of the code:

- A_c, B_c replaced by A_{cf}, B_{cf} ;
- R depends on A_c, B_c , replaced by S ;

Example System, 32-bit Floating-Point Numbers

```
xc = zeros(2, 1);  
%  $x_c \in \mathcal{E}_P$   
receive(y, 2); receive(yd, 3);  
%  $x_c \in \mathcal{E}_P$   
while (1)  
    %  $x_c \in \mathcal{E}_P$   
    yc = max(min(y - yd, 1), -1);  
    %  $x_c \in \mathcal{E}_P, y_c^2 \leq 1$   
    %  $\begin{pmatrix} x_c \\ y_c \end{pmatrix} \in \mathcal{E}_{Q_\mu}, Q_\mu = \begin{pmatrix} \mu^P & 0 \\ 0 & 1-\mu \end{pmatrix}$   
    u = Cc*xc + Dc*yc;  
    %  $\begin{pmatrix} x_c \\ y_c \end{pmatrix} \in \mathcal{E}_{Q_\mu}$   
    xc = Acf*xc + Bcf*yc;  
    %  $x_c \in \mathcal{E}_S, S = [(A_{cf} \ B_{cf})Q_\mu^{-1}(A_{cf} \ B_{cf})^T]^{-1}$   
    send(u, 1);  
    %  $x_c \in \mathcal{E}_S$   
    receive(y, 2); receive(yd, 3);  
    %  $x_c \in \mathcal{E}_S$   
    %  $x_c \in \mathcal{E}_P$   
end
```

In the rest of the code:

- A_c, B_c replaced by A_{cf}, B_{cf} ;
- R depends on A_c, B_c , replaced by S ;
- Check if $\mathcal{E}_S \subset \mathcal{E}_P$.

Example System, 32-bit Floating-Point Numbers

② Replace functions:

...

% $\begin{pmatrix} x_c \\ y_c \end{pmatrix} \in \mathcal{E}_{Q_\mu}$

$\mathbf{x}_c = \mathbf{A}_{cf} * \mathbf{x}_c + \mathbf{B}_{cf} * \mathbf{y}_c;$

% $\mathbf{x}_c \in \mathcal{E}_S$, $S = [(\mathbf{A}_{cf} \ \mathbf{B}_{cf}) \mathbf{Q}_\mu^{-1} (\mathbf{A}_{cf} \ \mathbf{B}_{cf})^T]^{-1}$

...

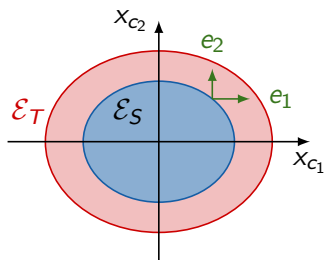
- Replace $+$ and \times by their FP counterparts;
- Increase \mathcal{E}_S to include arithmetic error.

Example System, 32-bit Floating-Point Numbers

e_1, e_2 is the arithmetic error on x_{c_1}, x_{c_2} .

$\mathcal{E}_T \supset \mathcal{E}_S$ is an ellipse s.t.:

$$\forall x_c \in \mathcal{E}_S, \forall x'_c \in \mathbb{R}^2, \\ |x'_{c_1} - x_{c_1}| \leq e_1 \wedge |x'_{c_2} - x_{c_2}| \leq e_2 \implies x'_c \in \mathcal{E}_T \quad (*)$$

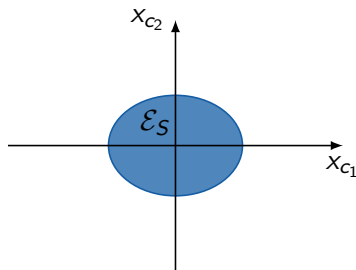


\mathcal{E}_T can be the smallest magnification of \mathcal{E}_S s.t. (*) holds.

Example System, 32-bit Floating-Point Numbers

```
...  
%  $\begin{pmatrix} x_c \\ y_c \end{pmatrix} \in \mathcal{E}_{Q_\mu}$   
xc = Acf*xc + Bcf*yc;  
%  $x_c \in \mathcal{E}_S$ ,  $S = [(A_{cf} \ B_{cf})Q_\mu^{-1}(A_{cf} \ B_{cf})^T]^{-1}$   
send(u, 1);  
%  $x_c \in \mathcal{E}_S$   
receive(y, 2); receive(yd, 3);  
%  $x_c \in \mathcal{E}_S$   
%  $x_c \in \mathcal{E}_P$   
end
```

In the rest of the code:

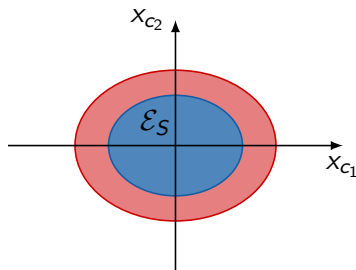


Example System, 32-bit Floating-Point Numbers

```
...  
%  $\begin{pmatrix} x_c \\ y_c \end{pmatrix} \in \mathcal{E}_{Q\mu}$   
xc = Acf*xc + Bcf*yc;  
%  $x_c \in \mathcal{E}_T$   
send(u, 1);  
%  $x_c \in \mathcal{E}_T$   
receive(y, 2); receive(yd, 3);  
%  $x_c \in \mathcal{E}_T$   
%  $x_c \in \mathcal{E}_P$   
end
```

In the rest of the code:

- Replace \mathcal{E}_S by \mathcal{E}_T ;



Example System, 32-bit Floating-Point Numbers

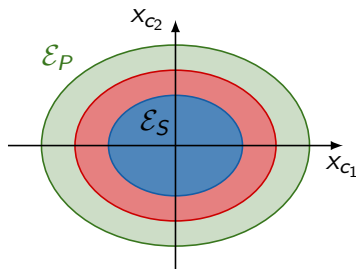
```
...  
%  $\begin{pmatrix} x_c \\ y_c \end{pmatrix} \in \mathcal{E}_{Q\mu}$   
xc = Acf*xc + Bcf*yc;  
%  $x_c \in \mathcal{E}_T$   
send(u, 1);  
%  $x_c \in \mathcal{E}_T$   
receive(y, 2); receive(yd, 3);  
%  $x_c \in \mathcal{E}_T$   
%  $x_c \in \mathcal{E}_P$   
end
```

In the rest of the code:

- Replace \mathcal{E}_S by \mathcal{E}_T ;
- Check if $\mathcal{E}_T \subset \mathcal{E}_P$.

It works! \Rightarrow Stable in 32 bits.

If not, can't conclude.



Automation: The LyaFloat Tool

In Python, using SymPy.

```
from lyafloat import *
setfloatify(constants=True, operators=True, precision=53)

P = Rational("1e-3") * Matrix(rationals(
    ["0.6742 0.0428", "0.0428 2.4651"]))
EP = Ellipsoid(P)
...
xc1, xc2, yc = symbols("xc1 xc2 yc")
Ac = Matrix(constants(["0.4990 -0.0500", "0.0100 1.0000"]))
...
ES = Ellipsoid(R)
print("ES included in EP :", ES <= EP)

i = Instruction({xc: Ac * xc + Bc * yc},
    pre=[zc in EQmu], post=[xc in ES])
ET = i.post()[xc]
print("ET =", ET)
print("ET included in EP :", ET <= EP)
```

Closed Loop

Closed-loop system:

- Pseudocode for controller and for environment;
- send & receive;
- Only controller code is changed.

Does not work with 32 bits.

OK with 128 bits.

Extensions of LyaFloat

Suitable method if bounded error.

① Arithmetic paradigms:

- OK with floating point: rounding error is bounded for $+$, $-$, $*$ if far enough from extremal values;
- Same for fixed point;
- Not sure what happens with two integers;

Extensions of LyaFloat

Suitable method if bounded error.

① Arithmetic paradigms:

- OK with floating point: rounding error is bounded for $+$, $-$, $*$ if far enough from extremal values;
- Same for fixed point;
- Not sure what happens with two integers;

② Other functions (non-linear systems):

- Differentiable, periodic functions (cos)
(can be computed with an abacus/polynomial interpolation);
- Differentiable functions restricted to a finite range
(assuming values in the range).

Related Work

Compute bounds from source code:

- Astrée;
- PhD P. Roux.

From pseudocode to C:

- Feron ICSM'10.

Floating-point arithmetic:

- PhD P. Roux.

Proof translation, code-level invariants.

Closed loop.

Conclusion

Theoretical framework to translate proof invariants on code with real arithmetic, while preserving the overall proof structure.

LyaFloat: implementation for Lyapunov-theoretic proofs on floating-point arithmetic.

Future work:

- Support for other arithmetic paradigms, more functions, more invariant propagators;
- Coq rather than Python
⇒ formalization (or proof?) of propagators;
- ...or generate Coq scripts?

Translation of Lyapunov Stability Proofs to Machine Arithmetic

Vivien Maisonneuve



Eighth meeting of the French community of compilation

Nice, July 2014