

# LEARN

## “A Lyapunov-enabled analysis of biochemical reaction networks”

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We describe the prerequisites of **LEARN**, the basic subroutines offered in **LEARN** and few example runs. We assume that the reader is familiar with [1].

## 1 Prerequisites

**LEARN** runs on MATLAB with the optimization and symbolic math toolboxes. Also, it needs the **cvx** package. The latest version of **cvx** is available on the link <http://cvxr.com/cvx/download/>. After download, the user must run **cvx\_setup**. After **cvx\_setup** reporting that **cvx** is successfully installed, **LEARN** should run without issues.

## 2 List of Subroutines

The following subroutines are available. Note that all the subroutines below take  $\Gamma$  as an input which is the stoichiometry matrix of the network. If the network has an autocatalytic reaction then both matrices  $A, B$  need to be entered. (see the Methods section in the main text)

### 2.1 Basic subroutines

- **LEARNmain(Gamma)**: Prints a basic report on the network. Examples will follow.
- **d=IsConservative(Gamma)**: Checks if the network is conservative. If it is, then the subroutine returns a positive vector  $d \in \mathbb{R}_+^n$  such that  $d^T \Gamma = 0$ . If the network is not conservative then  $d$  returns a scalar 0.
- **v=IsAS1(Gamma)**: Checks if the stoichiometry matrix has a positive vector in its kernel. If it does, then the subroutine returns a positive vector  $v \in \mathbb{R}_+^n$  such that  $\Gamma v = 0$ . If the network is not conservative then  $d$  returns a scalar 0.
- **[flag, deadlock]=checkSiphons(Gamma)**: Checks if there are critical siphons and deadlocks. Each output can be either 0 or 1.
- **flag=checkMnetwork(Gamma)**: Checks if the network is an  $M$ -network. The output is either 0 or 1.

## 2.2 Necessary Conditions

- `checkSiphonCondition(Gamma)`: Checks if the network violates the critical siphon necessary condition [1, Theorem 8]. It prints a brief report.
- `flag=SignPatternCheck(Gamma)`: Checks if the network violates the sign pattern necessary condition [2, Theorem 9]. The output is either 0 or 1.
- `flag=checkPmatreix(Gamma)`: Checks if the network violates the  $P$  matrix necessary condition [2, Theorem 8]. The output is either 0 or 1.
- `flag=RobustNondegeneracy(Gamma)`: Checks if the network has a robustly non-degenerate Jacobian [1, Theorem 7]. This only applies to networks that pass the  $P$  matrix test. The output is either 0 or 1.

## 2.3 Construction of RLFs

- `C=ConstructGraphical(Gamma)`: Checks if the network admits an RLF as given in Theorem 4. The output is  $C$ . If the method fails then  $C$  will be an empty matrix.
- `C=ConstructIterate(Gamma)`: Checks if the network admits an RLF as given in [1, Theorem 3]. The output is  $C$ . If the method fails then  $C$  will be an empty matrix.
- `[C, cvx]=ConstructLP(Gamma, H2, w, c)`: Checks if a non-autocatalytic network admits an RLF as given in [1, Theorem 2]. The last three inputs are optional. The output is  $C$  and the flag `cvx` to indicate whether the RLF is convex. The second input is  $H_2$  which are optional rows to add to the partitioning matrix  $H = \Gamma$ . The default value for  $H_2$  is an empty matrix. The third input is  $w$  and it is a flag to constrain the search to weighted  $\ell_1$  RLFs. The default value is 0. The fourth input is a flag to constrain the RLF to be convex. The default value is 0.
- `[C, cvx]=ConstructLPauto(A, B, H2, w, c)`: Checks if an *autocatalytic* network admits an RLF as given in [1, Theorem 2]. The remaining input structure is similar to the previous subroutine.
- `[C, cvx]=ConstructCoP(Gamma, H2, c)`: Checks if a non-autocatalytic network admits an RLF as given in [1, Theorem 5]. The last two inputs are optional. The output is  $C$  and a flag `cvx` to indicate whether the RLF is convex. The second input is  $H_2$  which are optional rows to add to the partitioning matrix  $H = \Gamma$ . The default value for  $H_2$  is an empty matrix. The third input is a flag to constrain the RLF to be convex. The default value is 0.

## 2.4 Checking a candidate RLF

- `flag=CheckRLF(Gamma, C)`: Checks if  $\tilde{V} = \max_k c_k^T r$  is an RLF for a non-autocatalytic network with the stoichiometry matrix  $\Gamma$ .

## 3 Examples

All the examples are included in the folder `examples`.

### 3.1 The double processive PTM cycle

This is the form of the input to LEARN for the network depicted in Fig. 9-b.

```
Gamma=[
-1      1      0      0      0      0      0      1;
-1      1      0      0      0      0      1      0;
1      -1     -1      0      0      0      0      0;
0      0      0     -1      1      0      1      0;
0      0      0     -1      1      0      0      1;
0      0      0      1     -1     -1      0      0;
0      0      1      0      0      0     -1      0;
0      0      0      0      0      1      0     -1];
LEARNmain(Gamma)
```

Note that the stoichiometry matrix  $\Gamma$  can be easily written from a list of reactions. The output of LEARN is as follows:

```
-----
Welcome to LEARN v1.0, May 2019
Developed by M. Ali Al-Radhawi malirdwi@{northeastern.edu,mit.edu,
    gmail.com}

LEARN tries to construct a Robust Lyapunov Function for a given
    reaction network.
-----
The network has 8 species and 8 reactions.
The stoichiometric space is 5-dimensional.
The network has a positive vector in the kernel of the
    stoichiometry matrix, i.e. it has the potential for positive
    steady states.
The network is conservative.
The network has no critical siphons. It is structurally persistent.
-----
LEARN will check some necessary conditions
Necessary Condition # 1 ....
The critical siphon necessary condition is satisfied.
Necessary Condition # 2 ....
The sign pattern necessary condition is satisfied.
Necessary Condition # 3 ....
The P matrix necessary condition is satisfied.
-----
LEARN will search for a PWL RLF
Method # 1: Graphical Method ..
This is an M-network. The graphical criteria will be checked

Success!! A PWL RLF has been found.
The following is always a Lyapunov function for any monotone
    kinetics:  $V(x)=|| C \cdot R(x) ||_{\infty}$ ,
where C is given as follows:
```

0	0	1	0	0	-1	0	0
0	0	1	0	0	0	-1	0
0	0	1	0	0	0	0	-1
-1	1	1	0	0	0	0	0
0	0	1	-1	1	0	0	0
0	0	0	0	0	1	-1	0
0	0	0	0	0	1	0	-1
-1	1	0	0	0	1	0	0
0	0	0	-1	1	1	0	0
0	0	0	0	0	0	1	-1
-1	1	0	0	0	0	1	0
0	0	0	-1	1	0	1	0
-1	1	0	0	0	0	0	1
0	0	0	-1	1	0	0	1
1	-1	0	-1	1	0	0	0

The robust non-degeneracy test is passed.

Since the network is conservative and with no critical siphons then the following holds:

There exists a unique positive globally asymptotically stable steady state in each stoichiometric class.

-----

Method # 2: Iterative Method ..

Success!! A PWL RLF has been found.

The following is always a Lyapunov function for any monotone kinetics:  $V(x) = || C \cdot R(x) ||_{\infty}$ ,

where C is given as follows:

-1	1	0	0	0	0	0	1
-1	1	0	0	0	0	1	0
1	-1	-1	0	0	0	0	0
0	0	0	-1	1	0	1	0
0	0	0	-1	1	0	0	1
0	0	0	1	-1	-1	0	0
0	0	1	0	0	0	-1	0
0	0	0	0	0	1	0	-1
0	0	0	0	0	0	-1	1
0	0	-1	0	0	0	0	1
-1	1	0	0	0	1	0	0
0	0	0	0	0	-1	1	0
0	0	1	-1	1	0	0	0
0	0	-1	0	0	1	0	0
-1	1	0	1	-1	0	0	0

The robust non-degeneracy test is passed.

Since the network is conservative and with no critical siphons then the following holds:

There exists a unique positive globally asymptotically stable steady state in each stoichiometric class.

```

-----
Method # 3: Linear Programming Method ..
The partition matrix H is set to the default choice H=the
stoichiometry matrix ..
This method for constructing a PWL RLF has failed.
THE END.

```

### 3.2 The double distributive PTM cycle

This is the output of LEARNmain for the network depicted in Fig. 9-d.

```

-----
Welcome to LEARN v1.0, May 2019
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gmail.com}

LEARN tries to construct a Robust Lyapunov Function for a given
reaction network.
-----
The network has 9 species and 12 reactions.
The stoichiometric space is 6-dimensional.
The network has a positive vector in the kernel of the
stoichiometry matrix, i.e. it has the potential for positive
steady states.
The network is conservative.
The network has no critical siphons. It is structurally persistent.
-----
LEARN will check some necessary conditions
Necessary Condition # 1 ....
The critical siphon necessary condition is satisfied.
Necessary Condition # 2 ....
The sign pattern necessary condition is satisfied.
Necessary Condition # 3 ....
The P matrix necessary condition is violated. A PWL RLF does not
exist
-----
LEARN will search for a PWL RLF
Method # 1: Graphical Method ..
This is not an M-network. Method # 1 is not applicable.

-----
Method # 2: Iterative Method ..
This method for constructing a PWL RLF has failed.
-----
Method # 3: Linear Programming Method ..
The partition matrix H is set to the default choice H=the
stoichiometry matrix ..
This method for constructing a PWL RLF has failed.

```

THE END.

### 3.3 The McKeithan Network

This is the output of LEARNmain for the network depicted in Fig. 11-a with  $N = 2$ .

```
-----
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    gmail.com}

LEARN tries to construct a Robust Lyapunov Function for a given
    reaction network.
-----
The network has 5 species and 6 reactions.
The stoichiometric space is 3-dimensional.
The network has a positive vector in the kernel of the
    stoichiometry matrix, i.e. it has the potential for positive
    steady states.
The network is conservative.
The network has no critical siphons. It is structurally persistent.
-----
LEARN will check some necessary conditions
Necessary Condition # 1 ....
The critical siphon necessary condition is satisfied.
Necessary Condition # 2 ....
The sign pattern necessary condition is satisfied.
Necessary Condition # 3 ....
The P matrix necessary condition is satisfied.
-----
LEARN will search for a PWL RLF
Method # 1: Graphical Method ..
This is not an M-network. Method # 1 is not applicable.

-----
Method # 2: Iterative Method ..
Success!! A PWL RLF has been found.
The following is always a Lyapunov function for any monotone
    kinetics:  $V(x) = || C * R(x) ||_{\infty}$ ,
where C is given as follows:
-1      1      0      0      1      1
-1      1      0      0      1      1
1       -1     -1      0      0      0
0        0      1     -1     -1      0
0        0      0      1      0     -1
0        0     -1      0      1      1
-1       1      1     -1      0      1
-1       1      0      1      1      0
```

The robust non-degeneracy test is passed.  
 Since the network is conservative and with no critical siphons then  
 the following holds:

There exists a unique positive globally asymptotically stable  
 steady state in each stoichiometric class.

-----

Method # 3: Linear Programming Method ..

The partition matrix H is set to the default choice H=the  
 stoichiometry matrix ..

Success!! A PWL RLF has been found.

The following is always a Lyapunov function for any monotone  
 kinetics:  $V(x) = || C * R(x) ||_{\infty}$ ,

where C is given as follows:

0	0	0	1	0	-1
0	0	1	-1	-1	0
0	0	1	-0	-1	-1
1	-1	-1	0	0	0
1	-1	-1	1	0	-1
1	-1	-0	-1	-1	0
1	-1	-0	-0	-1	-1

The robust non-degeneracy test is passed.

Since the network is conservative and with no critical siphons then  
 the following holds:

There exists a unique positive globally asymptotically stable  
 steady state in each stoichiometric class.

THE END.

## References

1. M. Ali Al-Radhawi, David Angeli, and Eduardo Sontag. A computational framework for a Lyapunov-enabled analysis of biochemical reaction networks. *bioRxiv:696716*, 2019.
2. M. Ali Al-Radhawi and David Angeli. New approach to the stability of chemical reaction networks: Piecewise linear in rates lyapunov functions. *IEEE Transactions on Automatic Control*, 61(1):76–89, 2016.