



## **PyPHS: An open source Python library dedicated to the generation of passive guaranteed simulation code for multiphysical (audio) systems**

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# Introduction

## Objective : Numerical simulation of multiphysical systems

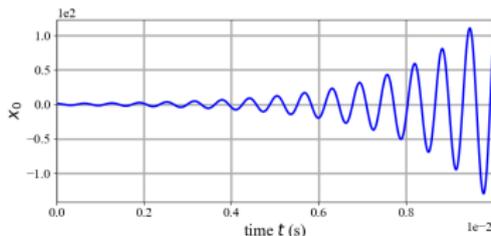
- electronics, mechanics, magnetics, thermics.
- nonlinearities, non ideal behaviors.
- high complexity.

## Standard approaches

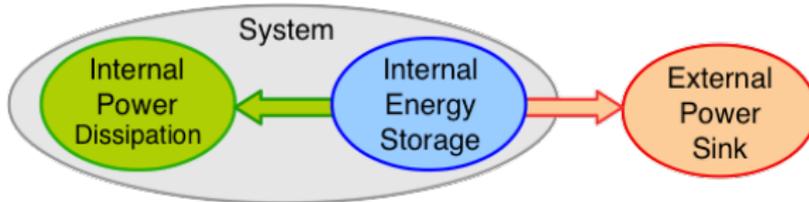
1. Build a set of elementary physical models.
2. Build a system as the connection of these models.
3. Apply *ad-hoc* discretization methods.

## Difficulties

- D1 The stability of a single model simulation is not guaranteed.  
D2 This is even worst for the interconnected system.



# But physical systems are passive systems



**Power-balance**  $\frac{dE}{dt} + P_D + P_S = 0$

with

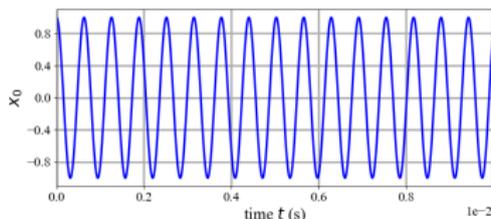
- Energy  $E$  (J),
- Dissipated power  $P_D$  (W),
- Sink Power  $P_S$  (W).

# Our approach

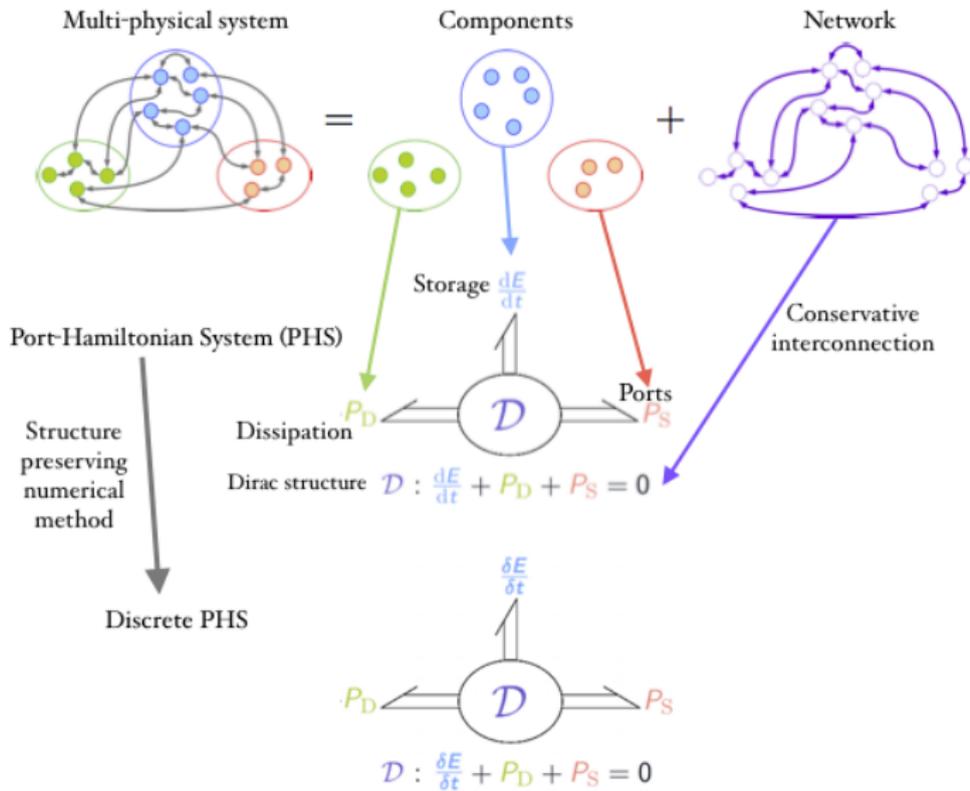
1. Structure physical models according to energy flows
2. Build a system as the structure preserving connection of these models
3. Apply a structure preserving discretization method

## Result

- D1 The stability of a single model simulation is guaranteed.
- D2 The interconnected system inherit from this property.



# Encoding of passivity in PyPHS



## Network are formal graph structures

- Use of NETWORKX<sup>1</sup> Python package.
- Creation and manipulation of graphs structures.

## Model equations in symbolic form

- Use of SYMPY<sup>2</sup> Python package.
- A posteriori manipulation of system's equations.
- Automated generation of L<sup>A</sup>T<sub>E</sub>X documentation.

## Numerical method is derived formally

- Also use SYMPY Python package.
- Symbolic optimization of the update equations.
- Easy analysis of the signal flow → Code generation.

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1. see <https://networkx.github.io/>

2. see <http://www.sympy.org/en/index.html>

## Main tools

- Port-Hamiltonian Systems (PHS) formalism<sup>3</sup>
- Graph theory<sup>4</sup>

## 2012→2016

- ANR project HaMecMoPSys<sup>5</sup>.
- PhD thesis of Antoine Falaize<sup>6</sup> in the team S3AM<sup>7</sup> at IRCAM - UMR STMS 9912 founded by EDITE.

## 2016→ . . .

- Implementation of the scientific results obtained between 2012 and 2016.
- Further scientific developments.

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3. MASCHKE, VAN DER SCHAFT et BREEDVELD, "An intrinsic Hamiltonian formulation of network dynamics : Non-standard Poisson structures and gyrators", 1992.

4. DESOER et KUH, Basic circuit theory, 2009.

5. see <https://hamecmopsys.ens2m.fr/>

6. FALAIZE, "Modélisation, simulation, génération de code et correction de systèmes multi-physiques audios :

## 1. Network

PyPHS inputs : Graph and Netlist.

## 1. Components

PyPHS dictionary elements : Graph objects.

## 3. Port-Hamiltonian Systems

PyPHS Core object : Passive-guaranteed structure.

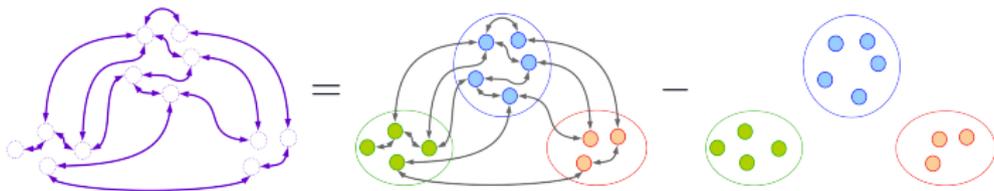
## 4. Numerical Method

PyPHS Method object : Structure preserving numerical scheme.

## 5. Code generation

PyPHS outputs : PYTHON, C++, JUCE and FAUST.

## Network



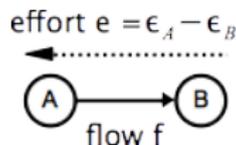
# System representation paradigm : Power graphs

## Directed graphs with self loops

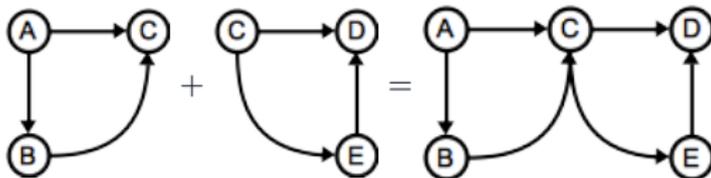
- Set of nodes  $N = \{N_1, \dots, N_n\}$  .
- Set of edges  $B = \{B_1, \dots, B_n\}$  with  $B_i = (n, m) \in N^2$ .
- Direction :  $B_i \equiv n \rightarrow m$

## Receiver convention

- Each edge  $\equiv$  two power variables : Flow and Effort
- Flow  $f$  : defined through the edges.
- Effort  $e$  : defined across the edges as the difference of two quantities.
- Power received by the edge :  $P = f e$  (W).



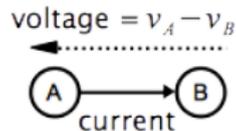
## Connection $\equiv$ Nodes identification



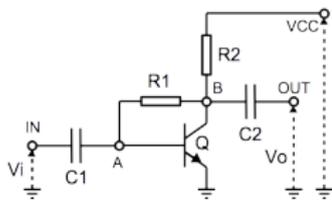
# Electrical graphs

## Physical quantities

Flow = Current (A), Effort = Voltage (V),  $\epsilon$  = Potential (V)

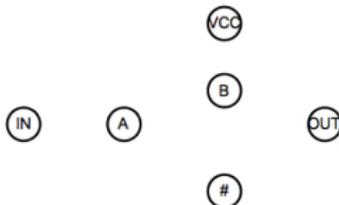


## Example system



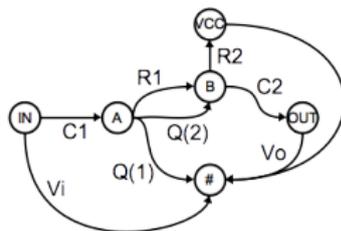
- 2 Capacitors C1 and C2,
- 2 Resistors R1 and R2,
- 1 BJ transistor Q,
- 3 Ports  $V_i$ ,  $V_o$  and  $V_c$ .

## Nodes



Graph nodes = Circuit nodes  
Ground = Reference node #

## Graph

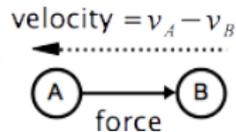


Graph edges = Circuit components  
Note  $Q \equiv 2$  edges

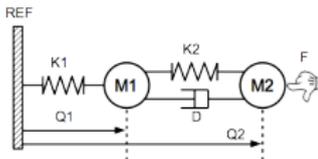
# Mechanical graphs

## Physical quantities

Flow = Force (N), Effort = Velocity (m/s),  $\epsilon$  = point velocity (m/s)



## Example system



- 2 Masses M1 and M2,
- 2 Springs K1 and K2,
- 1 Damper,
- 1 Port F.

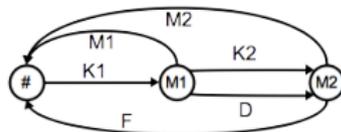
## Nodes



Graph nodes = unique velocities

Reference velocity = node #

## Graph

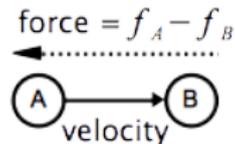


Graph edges = components

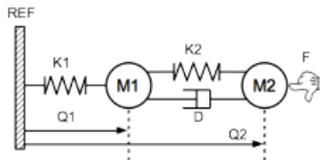
# Mechanical graphs (dual)

## Physical quantities

Flow = Velocity (m/s), Effort = force (N),  $\epsilon$  = some force (N)

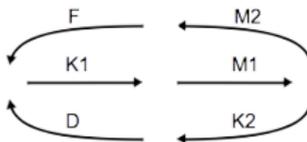


## Example system



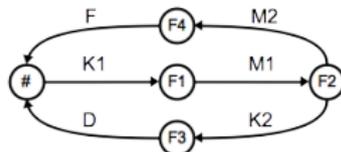
- 2 Masses M1 and M2,
- 2 Springs K1 and K2,
- 1 Damper,
- 1 Port F.

## Edges



Serial edges = same velocity

## Graph

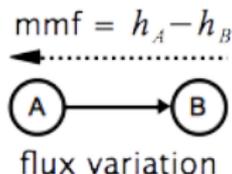


Add nodes to close the graph

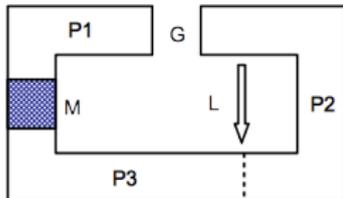
# Magnetical graphs

## Physical quantities

Flow = flux variation (V), Effort = magnetomotive force (A),  $\epsilon =$  some mmf (A)

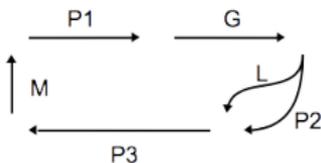


## Example system



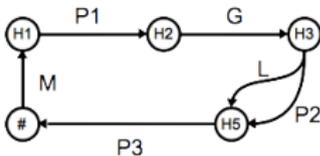
- 3 metal pieces P1, P2, P3,
- 1 Air gap G,
- 1 Flux leakage L,
- 1 Port M (magnet).

## Edges



Serial = same magnetic flux

## Graph

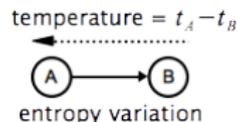


Add nodes to close the graph

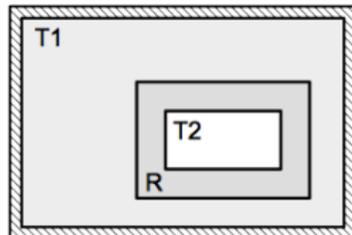
# Thermal graphs

## Physical quantities

Flow = entropy variation (W/K), Effort = temperature (K),  $\epsilon$  = temperature (K)



## Example system



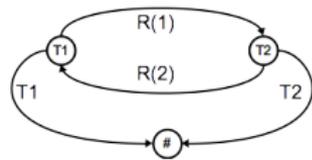
2 Heat capacities T1 and T2,  
1 Heat transfer R,

## Nodes



Graph nodes = temperatures  
Reference temperature = node #

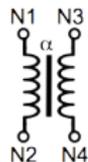
## Graph



Graph edges = components  
Note R = 2 edges (irreversibility)

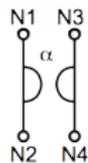
# Multiphysical graphs : connectors

## Transformer



$$\begin{aligned}e_{3 \rightarrow 4} &= \frac{1}{\alpha} e_{1 \rightarrow 2}, \\f_{3 \rightarrow 4} &= -\alpha f_{1 \rightarrow 2}, \\[\alpha] &= \frac{[f_{3 \rightarrow 4}]}{[f_{1 \rightarrow 2}]}.\end{aligned}$$

## Gyrator



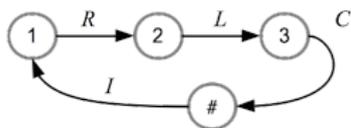
$$\begin{aligned}e_{3 \rightarrow 4} &= \alpha f_{1 \rightarrow 2}, \\f_{3 \rightarrow 4} &= -\frac{1}{\alpha} e_{1 \rightarrow 2}, \\[\alpha] &= \frac{[e_{3 \rightarrow 4}]}{[f_{1 \rightarrow 2}]}.\end{aligned}$$

## Conserving connection

In each case :  $P_{3 \rightarrow 4} = -P_{1 \rightarrow 2}$

# Kirchhoff laws on graphs

## Example : RLC



## Incidence Matrix

$$[\Gamma]_{n,b} = \begin{cases} 1 & \text{if edge } b \text{ is ingoing node } n, \\ -1 & \text{if edge } b \text{ is outgoing node } n. \end{cases}$$

$$\Gamma = \begin{array}{cccc|l} & B_R & B_L & B_C & B_I & \\ \hline & 0 & 0 & +1 & -1 & \# \\ N_1 & -1 & 0 & 0 & 0 & \\ N_2 & +1 & -1 & 0 & 0 & \\ N_3 & 0 & +1 & -1 & +1 & \end{array}$$

## Reduced incidence Matrix

Arbitrary reference node #

$$\Gamma = \begin{array}{cccc|l} & B_1 & \cdots & B_{n_B} & \\ \hline & \gamma_0 & & & \# \\ N_1 & & & & \\ \vdots & & & & \\ N_{n_N} & \gamma & & & \end{array},$$

## Generalized Kirchhoff's laws

- Efforts  $\mathbf{e} \in \mathbb{R}^{n_B}$ , flows  $\mathbf{f} \in \mathbb{R}^{n_N}$ .
- Node quantities  $\mathbf{p} \in \mathbb{R}^{n_N}$ .
- $\gamma^T \mathbf{p} = \mathbf{e}$ , (KVL).
- $\gamma \mathbf{f} = \mathbf{0}$ , (KCL).

# Dirac structure $\mathcal{D}$ = Kirchoff laws on graphs

## Edges splitting

Depends on the components

Flow controlled  $f \rightarrow \text{edge} \rightarrow e$ .

Effort controlled  $e \rightarrow \text{edge} \rightarrow f$ .

Outputs  $\mathbf{a} \in \mathbb{R}^{n_B}$ .

Inputs  $\mathbf{b} \in \mathbb{R}^{n_B}$ .

## Realizability criterion

If  $\gamma_f$  is invertible, then  $\exists! \mathbf{J}$  s.t.

$$\mathbf{b} = \mathbf{J} \cdot \mathbf{a}.$$

## Dirac structure

- $\mathbf{e}_b = \gamma_e^T \cdot \mathbf{p}$  and  $\mathbf{e}_a = \gamma_f^T \cdot \mathbf{p}$ ,
- $\gamma_e f_a = -\gamma_f \cdot f_b$ ,
- $\gamma_{ef} = \gamma_f^{-1} \cdot \gamma_e$ ,

## RLC example

$B_L$  is  $e$ -controlled,  $B_R, B_C, B_I$  are  $f$ -controlled.

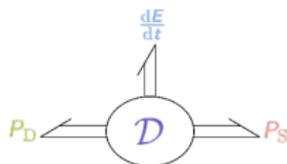
$$\left( \begin{array}{c|ccc} & B_L & B_R & B_C & B_I \\ \hline 0 & 0 & +1 & -1 \\ 0 & -1 & 0 & 0 \\ -1 & +1 & 0 & 0 \\ +1 & 0 & -1 & +1 \end{array} \right) \begin{array}{l} N_0 \\ N_1 \\ N_2 \\ N_3 \end{array}$$

$$\underbrace{\begin{pmatrix} e_b \\ f_b \end{pmatrix}}_{\mathbf{b}} = \underbrace{\begin{pmatrix} 0 & \gamma_{ef}^T \\ -\gamma_{ef} & 0 \end{pmatrix}}_{\mathbf{J}} \underbrace{\begin{pmatrix} f_a \\ e_b \end{pmatrix}}_{\mathbf{a}}$$

$\mathbf{J}$  is skew-symmetric  $\Rightarrow \mathbf{a}^T \cdot \mathbf{b} = \mathbf{a}^T \cdot \mathbf{J} \cdot \mathbf{a} = 0$ .

This is the Tellegen's theorem :

$$\sum_n^{n_B} e_n f_n = \sum_n^{n_B} P_n = 0.$$



## Automated construction of the Dirac structure

### Algorithme<sup>8</sup>

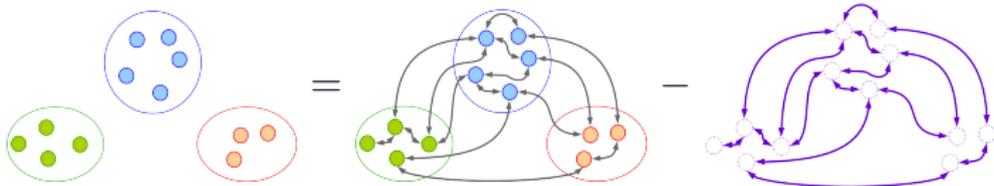
**Data** A **netlist** and a **dictionary** of components.

- Résult**
- If realizable :
    1. partition  $\mathbf{B} = [\mathbf{B}_e, \mathbf{B}_f]$ ,
    2. structure  $\mathbf{b} = \mathbf{J} \cdot \boldsymbol{\alpha}$ .
  - Else : Realizability fault detection  $\rightarrow$  the user correct the netlist.

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8. FALAIZE et HÉLIE, "Passive guaranteed simulation of analog audio circuits : A port-Hamiltonian approach", 2016.

## Components



# Storage components (definitions)

## Constitutive relation for component $s$

Storage function (Hamiltonian)  $H_s$  of the state  $x_s$ .

Stored energy  $E_s(t) = H_s(x_s(t)) \geq 0$ .

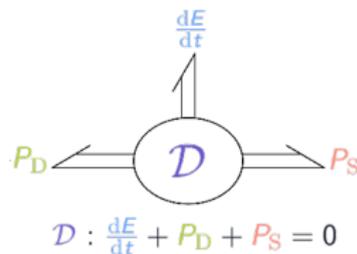
Received power  $\frac{dE_s}{dt} = H'_s(x_s) \frac{dx_s}{dt}$

## Power variables for component $s$

Received power  $\frac{dE_s}{dt} = e_s f_s$ .

$e$ -controlled  $e_s = \frac{dx_s}{dt} \implies f_s = H'_s(x_s)$ .

$f$ -controlled  $f_s = \frac{dx_s}{dt} \implies e_s = H'_s(x_s)$ .



## Total energy stored in $n_E$ storage edges

- $\mathbf{x} = (x_1, \dots, x_{n_E})$ .
- $E = H(\mathbf{x}) = \sum_{s=1}^{n_E} H_s(x_s) \geq 0$ .
- $\frac{dE}{dt} = \nabla H^T \frac{dx}{dt} = \sum_{s=1}^{n_E} \frac{dH_s}{dx_s} \frac{dx_s}{dt}$ .

## Storage components (examples)

### Mass (flow=velocity, effort=force)

**State** momentum  $x_m = m v_m$ .

**Hamiltonian** kinetic energy  $H_m(x_m) = \frac{x_m^2}{2m}$ .

**Flow** mass velocity  $f_m = H'_m(x_m) = \frac{x_m}{m}$ .

**Effort** inertial force  $e_m = \frac{dx_m}{dt} = m \frac{dv_m}{dt}$ .

### Capacitor

**State** charge  $q_C$ .

**Hamiltonian** electrostatic energy  $H_C(x_C) = \frac{x_C^2}{2C}$ .

**Flow** current  $f_C = \frac{dx_C}{dt} = \frac{dq_C}{dt}$ .

**Effort** voltage  $e_C = H'_C(x_C) = \frac{x_C}{C}$ .

# Dissipative components (definitions)

## Constitutive relation for component $d$

Dissipation function  $z_d$  of the variable  $w_d$ .

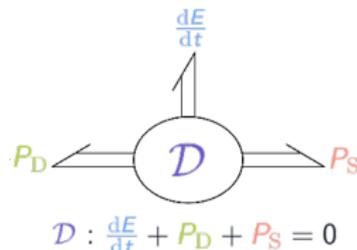
Received (dissipated) power  $P_{Dd}(t) = z_d(w_d(t)) \geq 0$ .

## Power variables for component $d$

Received power  $P_{Dd}(t) = e_d f_d \geq 0$

$e$ -controlled  $e_d = w_d \implies f_d = z_d(w_d)$ .

$f$ -controlled  $f_d = w_d \implies e_d = z_d(w_d)$ .



## Total power dissipated in $n_D$ dissipative edges

- $\mathbf{w} = (w_1, \dots, w_{n_D})$ .
- $\mathbf{z}(\mathbf{w}) = (z_1(w_1), \dots, z_{n_D}(w_{n_D}))$ .
- $P_D = \mathbf{z}(\mathbf{w})^T \cdot \mathbf{w} = \sum_{d=1}^{n_D} z_d(w_d) w_d \geq 0$ .

## Dissipative components (examples)

### Dashpot (flow=force, effort=velocity)

**Variable** elongation velocity  $w_D = v_D$ .

**Function** resistance force  $z_D(w_D) = D w_D$ , with  $D > 0$ .

**Flow** force  $f_D = z_D(w_D) = D v_D$ .

**Effort** velocity  $e_D = w_D = v_D$ .

Dissipated Power  $P_D = f_D e_D = R v_D^2$

### Resistor

**Variable** current  $w_R = i_R$ .

**Function** resistance voltage  $z_R(w_R) = R i_R$ , with  $R > 0$ .

**Flow** current  $f_R = w_R = i_R$ .

**Effort** voltage  $e_R = z_R(w_R) = R i_R$ .

Dissipated Power  $P_D = f_R e_R = R i_R^2$

# Ports (definitions)

## Input and output on port $p$

Actuated quantity  $u$  (input) and Observed quantity  $y$  (output).

Received Power  $P_{Sp}(t) = u_p(t) y_p(t)$ .

The power  $P_{Sp}$  is the power that goes out of the system on port  $p$ .

Ports are power sink.

## Power variables for port $p$

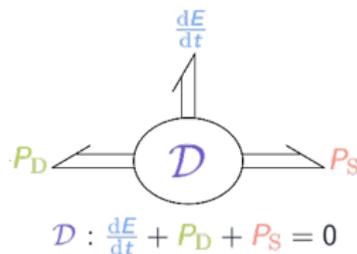
Received power  $P_{Sp}(t) = \epsilon_p f_p$

$\epsilon$ -controlled  $\epsilon_p = y_p \implies f_p = u_p$  (flow source).

$f$ -controlled  $f_p = y_p \implies \epsilon_p = u_p$  (effort source).

## Total power on $n_S$ port edges

- $\mathbf{u} = (u_1, \dots, u_{n_S})$ .
- $\mathbf{y} = (y_1, \dots, y_{n_S})$ .
- $P_S = \mathbf{u}^T \cdot \mathbf{y} = \sum_{p=1}^{n_S} u_p y_p$ .



### Voltage source

**Input** voltage  $u_U = v_U$ .

**Output** current  $y_U = i_U$ .

**Flow** current  $f_U = y_U$ .

**Effort** voltage  $e_U = u_U$ .

**Received Power**  $P_S = f_U e_U = v_U i_U$ .

### Imposed force (flow=force, effort=velocity)

**Input** force  $u_U = f_U$ .

**Output** velocity  $y_U = v_U$ .

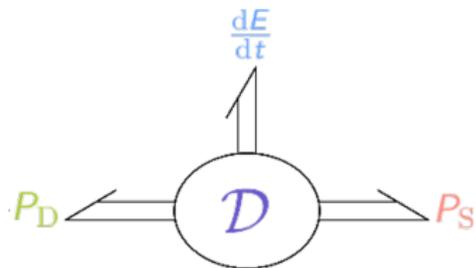
**Flow** force  $f_U = u_U$ .

**Effort** velocity  $e_U = y_U$ .

**Received Power**  $P_S = f_U e_U = f_U v_U$ .

- **Mechanics (1D)** : masses, springs lin./nonlin. (cubic, saturating, etc.), lin./nonlin. damping, visco-elastic (fractional derivatives).
- **Electronics** : batteries, coils and lin./nonlin. capacitors, resistors, transistors, diodes, triodes.
- **Magnetics** : Magnets, lin./nonlin capacitors, resisto-inductor (fractional integrators).
- **Thermics** : heat sources, capacitors.
- **Connections** : electromagnetic couplings, electromechanic coupling, irreversible transfers, gyrators, transformers.

### 3. Port-Hamiltonian Systems



$$\mathcal{D} : \frac{dE}{dt} + P_D + P_S = 0$$

# Putting all together

## Components

Storage  $\mathbf{b}_x = \frac{dx}{dt}$ ,  $\mathbf{a}_x = \nabla H(\mathbf{x})$

Dissipation  $\mathbf{b}_w = \mathbf{w}$ ,  $\mathbf{a}_w = \mathbf{z}(\mathbf{w})$

Ports  $\mathbf{b}_y = \mathbf{y}$ ,  $\mathbf{b}_y = \mathbf{u}$

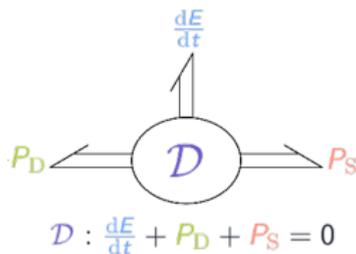
## Network (Dirac structure)

$$\mathbf{b} = \begin{pmatrix} \mathbf{b}_x \\ \mathbf{b}_w \\ \mathbf{b}_y \end{pmatrix} \text{ and } \mathbf{a} = \begin{pmatrix} \mathbf{a}_x \\ \mathbf{a}_w \\ \mathbf{a}_y \end{pmatrix}$$

with  $\mathbf{b} = \mathbf{J} \cdot \mathbf{a}$  and  $\mathbf{J}^T = -\mathbf{J}$ .

## This encodes the power balance

$$\begin{aligned} 0 &= \mathbf{a}^T \cdot \mathbf{b} \\ &= \underbrace{\nabla H(\mathbf{x})^T \cdot \frac{dx}{dt}}_{\frac{dE}{dt}} + \underbrace{\mathbf{z}(\mathbf{w}) \cdot \mathbf{w}}_{P_D} + \underbrace{\mathbf{u}^T \cdot \mathbf{y}}_{P_S} \end{aligned}$$



# Port-Hamiltonian structure

$$\underbrace{\begin{pmatrix} \text{Storage} \\ \text{Dissipation} \\ \text{Ports} \end{pmatrix} \begin{pmatrix} \frac{dx}{dt} \\ \mathbf{w} \\ \mathbf{y} \end{pmatrix}}_{\mathbf{b}} = \underbrace{\begin{pmatrix} +\mathbf{J}_{xx} & +\mathbf{J}_{xw} & +\mathbf{J}_{xy} \\ -\mathbf{J}_{xw}^T & +\mathbf{J}_{ww} & +\mathbf{J}_{wy} \\ -\mathbf{J}_{xy}^T & -\mathbf{J}_{wy}^T & +\mathbf{J}_{yy} \end{pmatrix}}_{\mathbf{J}} \cdot \underbrace{\begin{pmatrix} \nabla H(\mathbf{x}) \\ \mathbf{z}(\mathbf{w}) \\ \mathbf{u} \end{pmatrix}}_{\mathbf{a}}$$

# Reduction of the linear dissipative structure<sup>9</sup>

## Splitting of $\mathbf{z}$

$\mathbf{Z}_1$  a diagonal matrix and  $\mathbf{z}_{nl}$  a collection of nonlinear functions

$$\mathbf{w} = \begin{pmatrix} \mathbf{w}_1 \\ \mathbf{w}_{nl} \end{pmatrix}, \quad \mathbf{z}(\mathbf{w}) = \begin{pmatrix} \mathbf{Z}_1 \cdot \mathbf{w}_1 \\ \mathbf{z}_{nl}(\mathbf{w}_{nl}) \end{pmatrix},$$

## New Port-Hamiltonian structure

$$\underbrace{\begin{pmatrix} \frac{dx}{dt} \\ \mathbf{w}_{nl} \\ \mathbf{y} \end{pmatrix}}_{\hat{\mathbf{b}}} = \underbrace{(\hat{\mathbf{J}} - \mathbf{R})}_{\mathbf{M}} \cdot \underbrace{\begin{pmatrix} \nabla H(\mathbf{x}) \\ \mathbf{z}_{nl}(\mathbf{w}_{nl}) \\ \mathbf{u} \end{pmatrix}}_{\hat{\mathbf{a}}}$$

## Interpretation

- $\hat{\mathbf{J}}$   $\rightarrow$  reduced conservative interconnection,
- $\mathbf{R} \succeq 0 \rightarrow$  resistive interconnection (includes the coefficients from  $\mathbf{Z}_1$ ).

9. FALAIZE et HÉLIE, "Passive guaranteed simulation of analog audio circuits : A port-Hamiltonian approach", 2016.

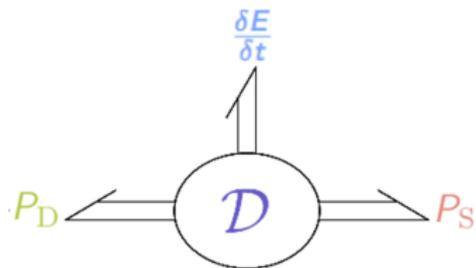
# PyPHS Port-Hamiltonian structure

$$\underbrace{\begin{pmatrix} \frac{dx}{dt} \\ \mathbf{w} \\ \mathbf{y} \end{pmatrix}}_{\mathbf{b}} = \underbrace{\begin{pmatrix} M_{xx} & M_{xw} & M_{xy} \\ M_{wx} & M_{ww} & M_{wy} \\ M_{yx} & M_{yw} & M_{yy} \end{pmatrix}}_{\mathbf{M}} \cdot \underbrace{\begin{pmatrix} \nabla H(\mathbf{x}) \\ \mathbf{z}(\mathbf{w}) \\ \mathbf{u} \end{pmatrix}}_{\mathbf{a}}$$

with

$$\mathbf{M} = \underbrace{\begin{pmatrix} +\mathbf{J}_{xx} & +\mathbf{J}_{xw} & +\mathbf{J}_{xy} \\ -\mathbf{J}_{xw}^T & +\mathbf{J}_{ww} & +\mathbf{J}_{wy} \\ -\mathbf{J}_{xy}^T & -\mathbf{J}_{wy}^T & +\mathbf{J}_{yy} \end{pmatrix}}_{\mathbf{J}} - \underbrace{\begin{pmatrix} \mathbf{R}_{xx} & \mathbf{R}_{xw} & \mathbf{R}_{xy} \\ \mathbf{R}_{xw}^T & \mathbf{R}_{ww} & \mathbf{R}_{wy} \\ \mathbf{R}_{xy}^T & \mathbf{R}_{wy}^T & \mathbf{R}_{yy} \end{pmatrix}}_{\mathbf{R}}$$

## 4. Numerical method



$$\mathcal{D} : \frac{\delta E}{\delta t} + P_D + P_S = 0$$

# Structure preserving numerical method 1

## Objective

Discrete time power balance :  $\frac{\delta E}{\delta T}[k] + P_D[k] + P_S[k] = 0$ .

## Choice

- $\frac{\delta E[k]}{\delta T} = \frac{E[k+1]-E[k]}{\delta T} = \frac{H(\mathbf{x}[k+1])-H(\mathbf{x}[k])}{\delta T}$
- Mono-variate case :

$$\frac{E[k+1] - E[k]}{\delta T} = \sum_n \frac{H_n(x_n[k+1]) - H_n(x_n[k])}{x_n[k+1] - x_n[k]} \cdot \frac{x_n[k+1] - x_n[k]}{\delta T}$$

## Solution :

$$\begin{aligned} \frac{dx}{dt} &\longrightarrow \frac{\delta \mathbf{x}[k]}{\delta T} = \frac{\mathbf{x}[k+1]-\mathbf{x}[k]}{\delta T} \\ \nabla H(\mathbf{x}) &\longrightarrow \nabla^d H(\mathbf{x}[k], \delta \mathbf{x}[k]) \triangleq \text{discrete gradient}^{10} \end{aligned}$$

with

$$\left[ \nabla^d H(\mathbf{x}, \delta \mathbf{x}) \right]_n = \frac{H_n([\mathbf{x} + \delta \mathbf{x}]_n) - H_n([\mathbf{x}]_n)}{[\delta \mathbf{x}]_n} \xrightarrow{[\delta \mathbf{x}]_n \rightarrow 0} \frac{dH_n}{dx_n}(x_n).$$

10. ITOH et ABE, "Hamiltonian-conserving discrete canonical equations based on variational difference quotients",

## Solution

$$\begin{aligned} \frac{dx}{dt} &\longrightarrow \frac{\delta x[k]}{\delta T} = \frac{x[k+1]-x[k]}{\delta T} \\ \nabla H(\mathbf{x}) &\longrightarrow \nabla^d H(\mathbf{x}[k], \delta \mathbf{x}[k]) \end{aligned}$$

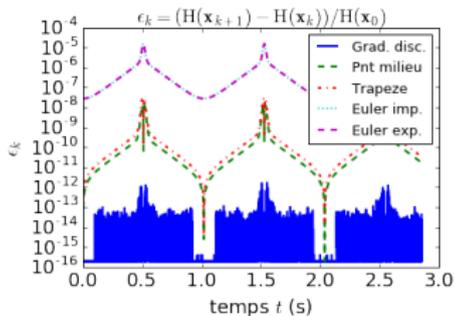
## Discret PHS

$$\begin{pmatrix} \frac{\delta x[k]}{\delta T} \\ \mathbf{w}[k] \\ \mathbf{y}[k] \end{pmatrix} = \mathbf{M} \cdot \begin{pmatrix} \nabla^d H(\mathbf{x}[k], \delta \mathbf{x}[k]) \\ \mathbf{z}(\mathbf{w}[k]) \\ \mathbf{u}[k] \end{pmatrix}.$$

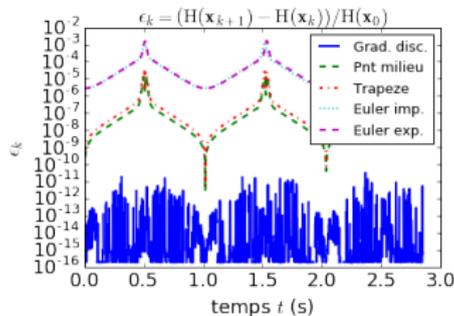
PHS structure is preserved in discrete time  $\Rightarrow$  numerical passivity.

# Relative error on the power balance (PyPHS in blue)

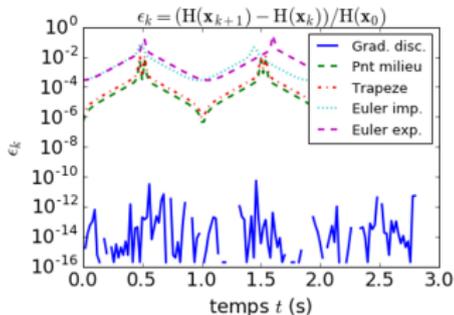
$f_e = 5000\text{Hz}$



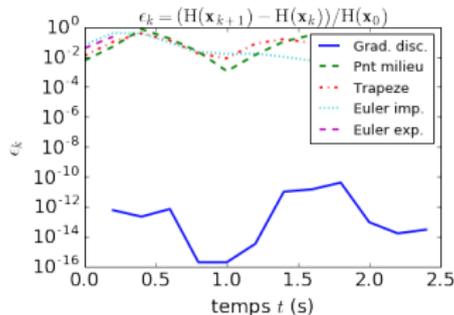
$f_e = 500\text{Hz}$



$f_e = 50\text{Hz}$

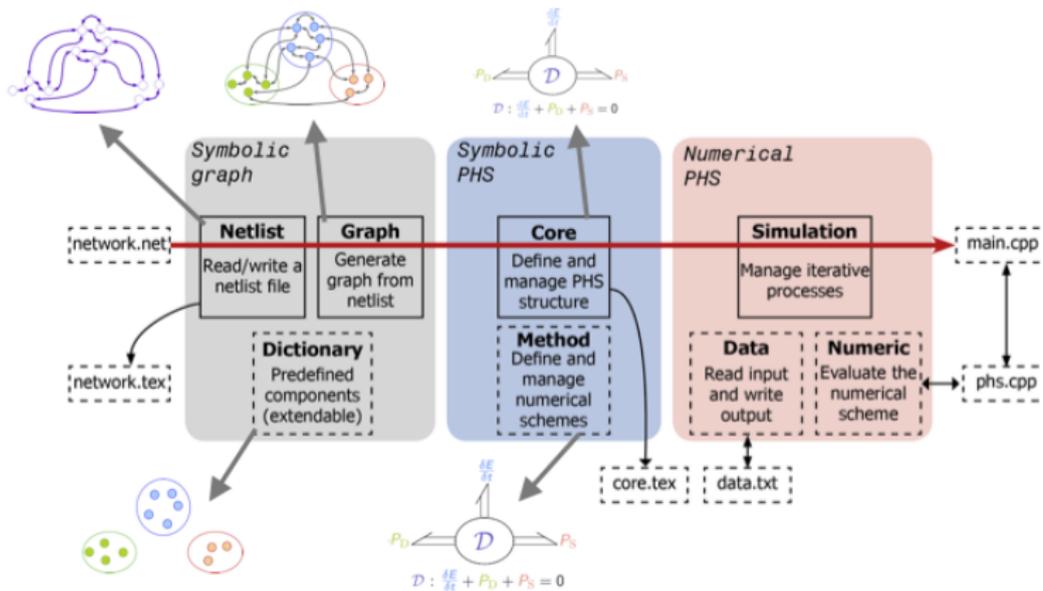


$f_e = 5\text{Hz}$



## 5. Code generation

# PyPHS : an overview



## Formal Method **object to numerical Simulation object**

1. Parameters are substituted in the discrete PHS.
2. Each symbolic expression is simplified and transformed into Python functions.
3. Updates of internal variable is defined by a message passing system.

## Perform simulation

- Inputs are :
  1. A sequence of input values,
  2. A sequence of control parameters values.
- Apply each update sequentially.
- Results are stored on disk to avoid memory overload.

## Formal Method object to C++ code

1. Parameters are associated to pointers → can be changed after generation.
2. Each symbolic expression is simplified and transformed into a C++ function.
3. Same message passing system.

## Perform simulation

- Inputs are :
  1. the sample rate,
  2. a sequence of input values,
  3. a sequence of control parameters values.
- Apply each update sequentially.
- Results are stored on disk → call back into Python for post processing.

## Only for Juce audio FX

1. Call the generated C++ object into Juce Template.
2. Generation of a set of snippets → copy/past into Juce template.
3. The control parameters are automatically associate with sliders → real-time control.
4. Still experimental.

## Yield AU/VST real-time audio plugins

- Can be used in most Digital Audio Workstations.

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11. <https://juce.com/>

## Only for FAUST audio FX

- Dedicated Method object : Symbolic pre-inversion of every matrices.
- Fixed number of nonlinear solvers iteration → duplicate of a single iteration.
- A complete iteration is built and encompassed in a dedicated feedback system.
- Control parameters are associated with sliders.
- Still experimental.

## Yield VST real-time audio plugins

- Automated optimization of the signal flow.
- Can be used in most Digital Audio Workstations.
- Several compilation targets.

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12. <http://faust.grame.fr/>

**Last word**

- Open source Library on a GITHUB repository<sup>13</sup>.
- Licence CECILL (CEA-CNRS-INRIA Logiciels libres).
- PYTHON 2.7 & 3.5 supported, Mac OSX, Windows 10 and Linux.
- Multiphysical components dictionary.
- Automated graph analysis.
- Automated derivation of the PHS structure and L<sup>A</sup>T<sub>E</sub>X code generation.
- Passive guaranteed simulations.
- Automated generation of C++, JUCE and FAUST code.

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13. <https://pyphs.github.io/pyphs/>

## Scientific results to be implemented

- Anti-aliasing observer (PhD Remy Müller).
- PHS in scattering variables ( $\rightsquigarrow$  Wave Digital PHS).
- Piecewise Linear constitutive laws ( $\rightsquigarrow$  cope with realizability faults).
- Improve Nonlinear solver ( $\neq$  Newton-Raphson).
- Automated derivation of a command laws (feedforward + feedback).
- ...

## **Accelerate development**

CALL FOR DEVELOPERS

## **Improve robustness**

CALL FOR USERS

**Thank you for your attention**

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